

Advanced topics in Markov-chain Monte Carlo

Lecture 1:

Transition matrices - from the balance conditions to mixing

Part 1/2: Introduction

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References

- W. Krauth “**Statistical Mechanics: Algorithms and Computations**” (Oxford University Press, 2006; second edition: “soon”)

Monte Carlo



Direct sampling (algorithm)

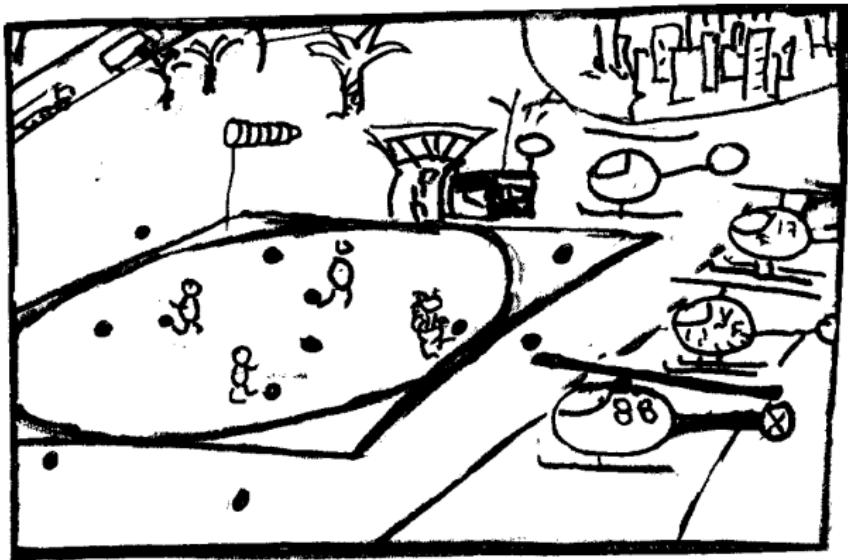
```
procedure direct-pi
     $N_{\text{hits}} \leftarrow 0$  (initialize)
    for  $i = 1, \dots, N$  do
         $\begin{cases} x \leftarrow \text{ran}[-1, 1] \\ y \leftarrow \text{ran}[-1, 1] \\ \text{if } (x^2 + y^2 < 1) \quad N_{\text{hits}} \leftarrow N_{\text{hits}} + 1 \end{cases}$ 
    output  $N_{\text{hits}}$ 
```

Direct sampling (results)

Five trials with $N = 4000$

run	N_{hits}	estimation
1	3156	3.156
2	3129	3.129
3	3154	3.154
4	3134	3.134
5	3148	3.148

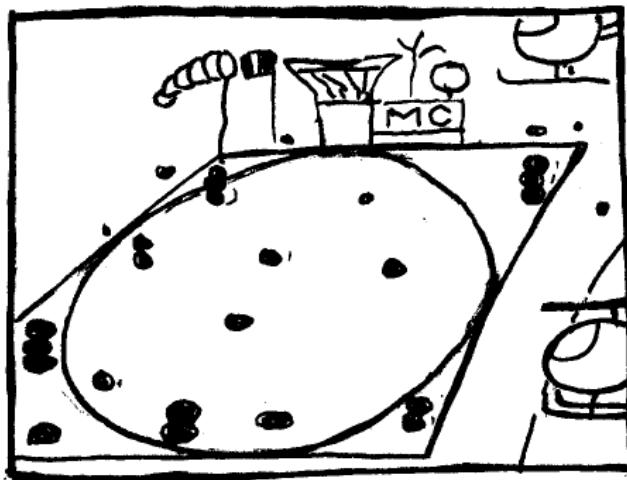
Markov-chain sampling (1/3)



Markov-chain sampling (2/3)

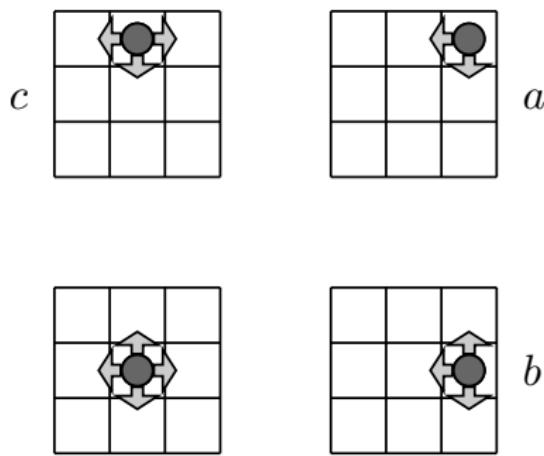
```
procedure markov-pi
    hits  $\leftarrow$  0;  $x \leftarrow 1$ ;  $y \leftarrow -1$ 
    for  $i = 1, \dots, N$  do
         $\left\{ \begin{array}{l} \delta x \leftarrow \text{ran}[-\delta, \delta] \\ \delta y \leftarrow \text{ran}[-\delta, \delta] \\ \text{if } (|x + \delta x| < 1 \text{ and } |y + \delta y| < 1) \text{ then} \\ \quad \left\{ \begin{array}{l} x \leftarrow x + \delta x \\ y \leftarrow y + \delta y \end{array} \right. \\ \text{if } (x^2 + y^2 < 1) N_{\text{hits}} \leftarrow N_{\text{hits}} + 1 \end{array} \right.$ 
    output  $N_{\text{hits}}$ 
```

Markov-chain sampling (3/3)



- Metropolis et al. (1953).

3×3 pebble game



- discretized version of heliport game

Detailed balance

$$\underbrace{p(a \rightarrow a)}_{\text{probability to go from } a \text{ to } a} + p(a \rightarrow b) + p(a \rightarrow c) = 1$$

$$\underbrace{\pi(a)}_{\text{probability to be at } a} = \pi(a)p(a \rightarrow a) + \pi(b)p(b \rightarrow a) + \pi(c)p(c \rightarrow a)$$

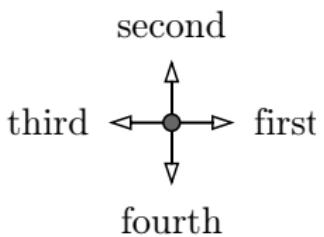
$$\pi(a)p(a \rightarrow c) + \pi(a)p(a \rightarrow b) = \pi(b)p(b \rightarrow a) + \pi(c)p(c \rightarrow a)$$

detailed balance condition

$$\pi(a)p(a \rightarrow b) = \pi(b)p(b \rightarrow a) \quad \text{etc}$$

A priori probabilities, acceptance, implementation

7	8	9
4	5	6
1	2	3



site k	Nbr(., k)			
	1	2	3	4
1	2	4	0	0
2	3	5	1	0
3	0	6	2	0
4	5	7	0	1
5	6	8	4	2
6	0	9	5	3
7	8	0	0	4
8	9	0	7	5
9	0	0	8	6

Transition matrix

- P : matrix of conditional transition probabilities from i to j :

$$P = \begin{bmatrix} \frac{1}{2} & \frac{1}{4} & \cdot & \frac{1}{4} & \cdot & \cdot & \cdot & \cdot & \cdot \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \cdot & \frac{1}{4} & \cdot & \cdot & \cdot & \cdot \\ \cdot & \frac{1}{4} & \frac{1}{2} & \cdot & \cdot & \frac{1}{4} & \cdot & \cdot & \cdot \\ \frac{1}{4} & \cdot & \cdot & \frac{1}{4} & \frac{1}{4} & \cdot & \frac{1}{4} & \cdot & \cdot \\ \cdot & \frac{1}{4} & \cdot & \frac{1}{4} & 0 & \frac{1}{4} & \cdot & \frac{1}{4} & \cdot \\ \cdot & \cdot & \frac{1}{4} & \cdot & \frac{1}{4} & \frac{1}{4} & \cdot & \cdot & \frac{1}{4} \\ \cdot & \cdot & \cdot & \frac{1}{4} & \cdot & \cdot & \frac{1}{2} & \frac{1}{4} & \cdot \\ \cdot & \cdot & \cdot & \cdot & \frac{1}{4} & \cdot & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \\ \cdot & \cdot & \cdot & \cdot & \cdot & \frac{1}{4} & \cdot & \frac{1}{4} & \frac{1}{2} \end{bmatrix}$$

Algorithmic probabilities II

- initial probability vector

$$\pi^{\{t=0\}} = \{0, \dots, 0, 1\}.$$

- probability at iteration $i + 1$ from iteration i

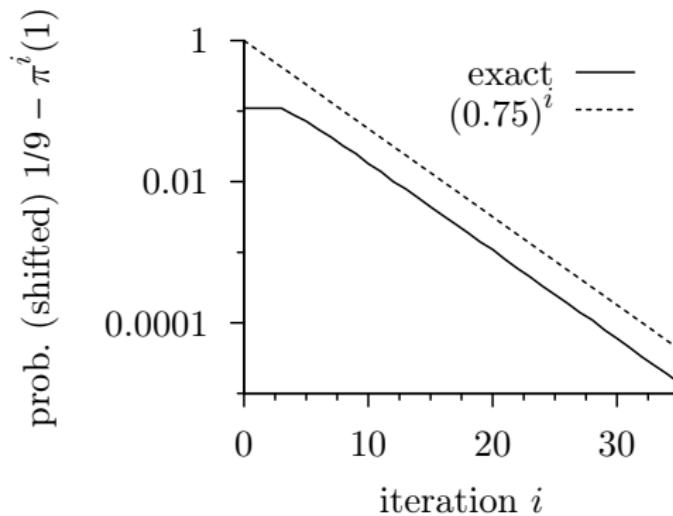
$$\pi_i^{\{t+1\}} = \sum_{j=1}^9 \pi_j^{\{t\}} P_{j \rightarrow i}$$

- eigenvectors and eigenvalues

$$\{\pi_1^{(t)}, \dots, \pi_9^{(t)}\} = \underbrace{\left\{\frac{1}{9}, \dots, \frac{1}{9}\right\}}_{\begin{array}{l} \text{first eigenvector} \\ \text{eigenvalue } \lambda_1 = 1 \end{array}} + \alpha_2 (0.75)^t \underbrace{\{-0.21, \dots, 0.21\}}_{\begin{array}{l} \text{second eigenvector} \\ \text{eigenvalue } \lambda_2 = 0.75 \end{array}} + \dots$$

Transition matrices

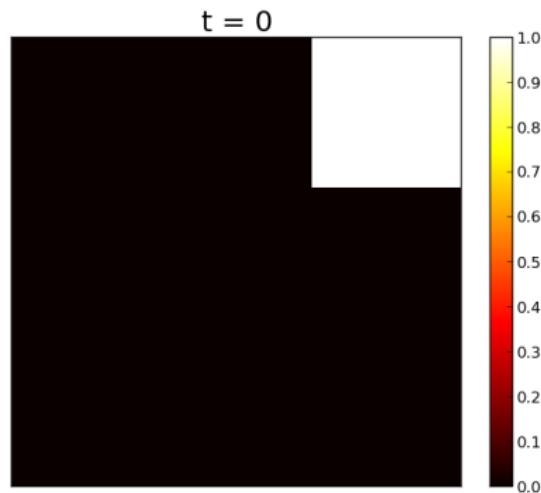
- convergence of pebble game:



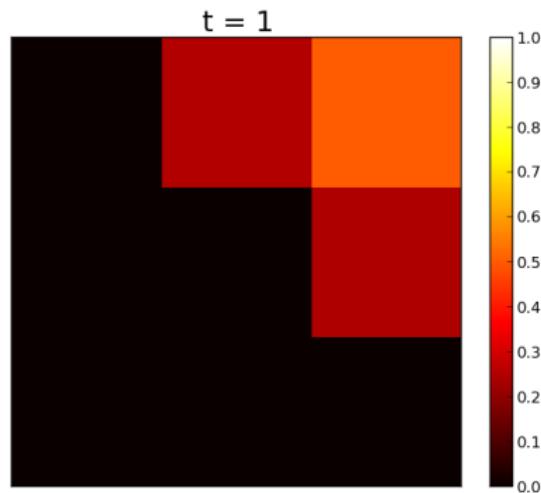
- exponential convergence \equiv scale

$$(0.75)^t = \exp [t \cdot \log 0.75] = \exp \left[-\frac{t}{3.476} \right].$$

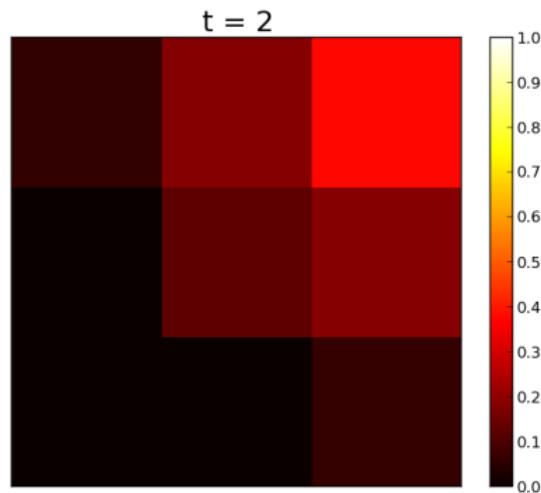
- Distribution $\pi^{t=0}$ (starting from upper right)



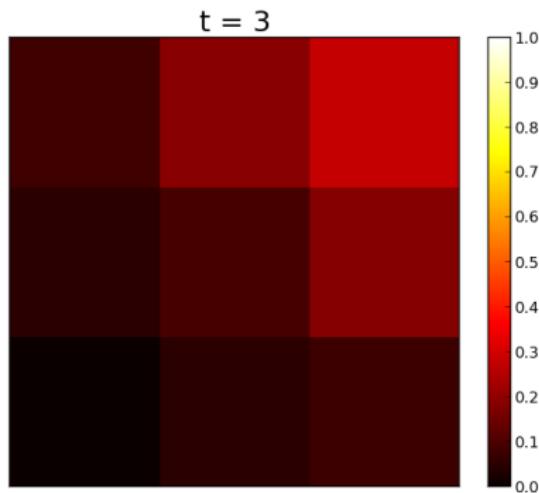
- Distribution $\pi^{t=1}$ (starting from upper right)



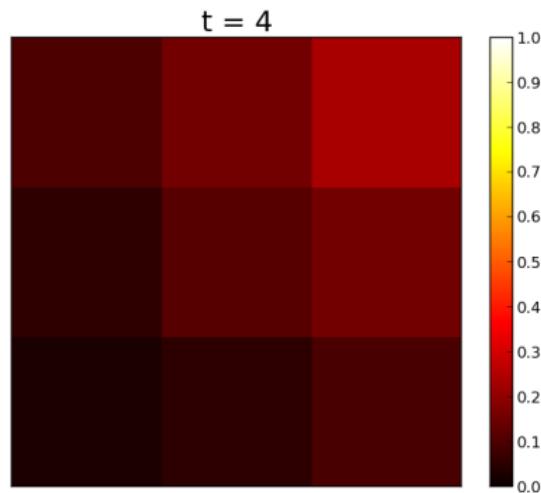
- Distribution $\pi^{t=2}$ (starting from upper right)



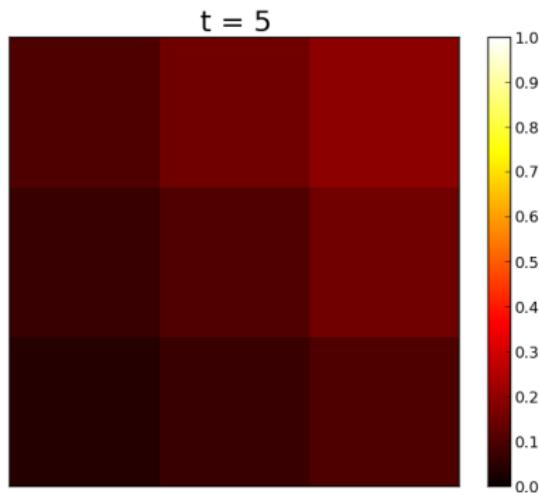
- Distribution $\pi^{t=3}$ (starting from upper right)



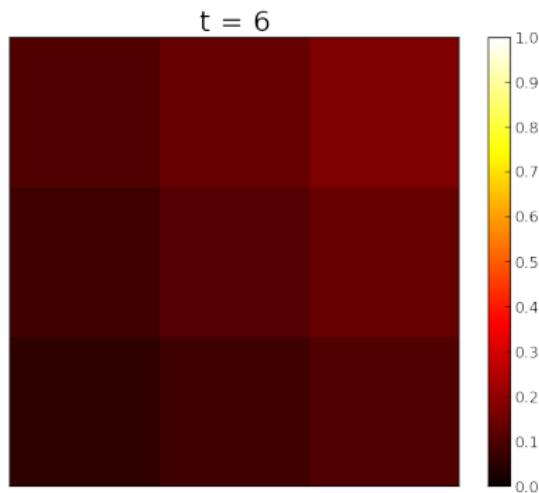
- Distribution $\pi^{t=4}$ (starting from upper right)



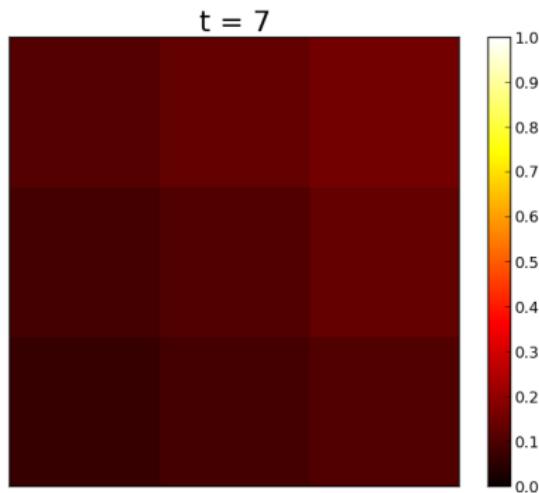
- Distribution $\pi^{t=5}$ (starting from upper right)



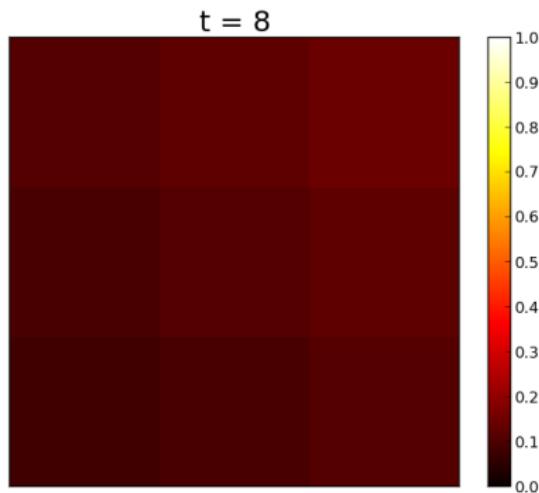
- Distribution $\pi^{t=6}$ (starting from upper right)



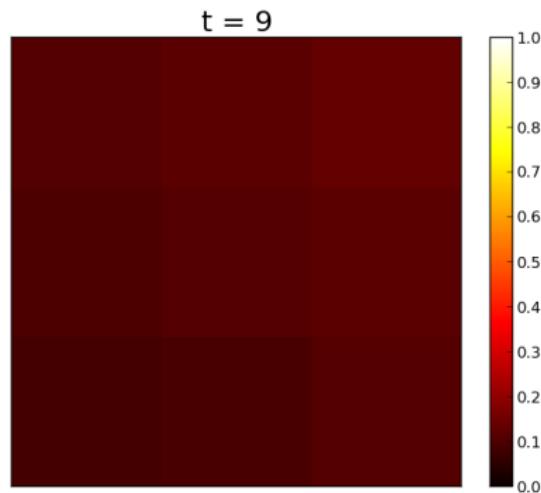
- Distribution $\pi^{t=7}$ (starting from upper right)



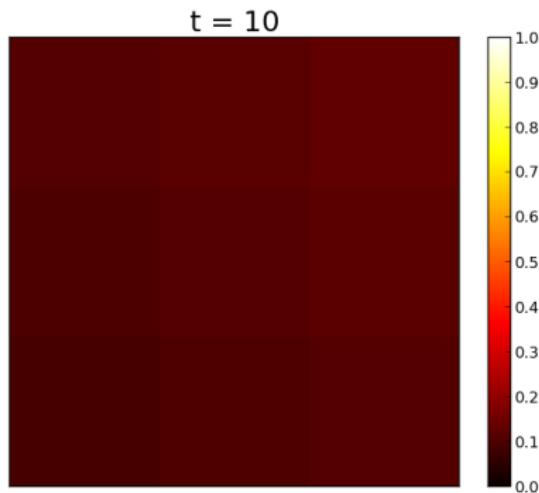
- Distribution $\pi^{t=8}$ (starting from upper right)



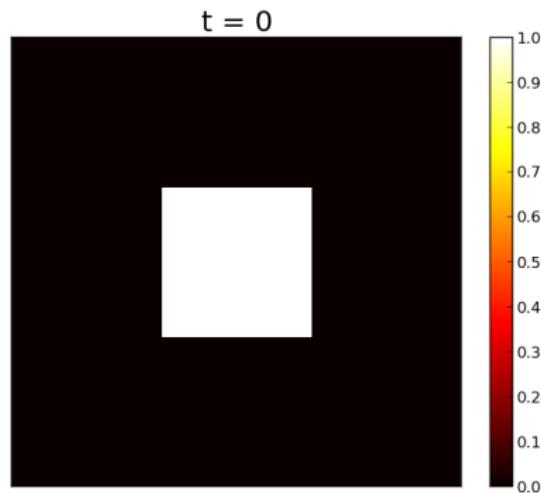
- Distribution $\pi^{t=9}$ (starting from upper right)



- Distribution $\pi^{t=10}$ (starting from upper right)



- Distribution $\pi^{t=0}$ (starting from center)



- Distribution $\pi^{t=1}$ (starting from center)

